

```
In [3]: ▶ import pandas as pd
df = pd.read_csv('C:\\Users\\jagad\\Downloads\\01.Data Cleaning and Preprocessing.csv')
print("First few rows of the DataFrame:")
print(df.head())
```

First few rows of the DataFrame:

	Observation	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	\
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	

  

	T-upperExt-2	T-lowerExt-2	UCZAA	WhiteFlow-4	...	SteamFlow-4	\
0	358.282	329.545	1.443	599.253	...	67.122	
1	351.050	329.067	1.549	537.201	...	60.012	
2	350.022	329.260	1.600	549.611	...	61.304	
3	350.938	331.142	1.604	623.362	...	68.496	
4	351.640	332.709	NaN	638.672	...	70.022	

  

	Lower-HeatT-3	Upper-HeatT-3	ChipMass-4	WeakLiquorF	BlackFlow-2	\
0	329.432	303.099	175.964	1127.197	1319.039	
1	330.823	304.879	163.202	665.975	1297.317	
2	329.140	303.383	164.013	677.534	1327.072	
3	328.875	302.254	181.487	767.853	1324.461	
4	328.352	300.954	183.929	888.448	1343.424	

  

	WeakWashF	SteamHeatF-3	T-Top-Chips-4	SulphidityL-4
0	257.325	54.612	252.077	NaN
1	241.182	46.603	251.406	29.11
2	237.272	51.795	251.335	NaN
3	239.478	54.846	250.312	29.02
4	215.372	54.186	249.916	29.01

[5 rows x 23 columns]

```
In [24]: ► filtered_df = df[df['BlowFlow'] > 10]
print("\nFiltered DataFrame:")
print(filtered_df.head())
```

Filtered DataFrame:

	Observation	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	\
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	

  

	T-upperExt-2	T-lowerExt-2	UCZAA	WhiteFlow-4	...	SteamFlow-4	\
0	358.282	329.545	1.443	599.253	...	67.122	
1	351.050	329.067	1.549	537.201	...	60.012	
2	350.022	329.260	1.600	549.611	...	61.304	
3	350.938	331.142	1.604	623.362	...	68.496	
4	351.640	332.709	NaN	638.672	...	70.022	

  

	Lower-HeatT-3	Upper-HeatT-3	ChipMass-4	WeakLiquorF	BlackFlow-2	\
0	329.432	303.099	175.964	1127.197	1319.039	
1	330.823	304.879	163.202	665.975	1297.317	
2	329.140	303.383	164.013	677.534	1327.072	
3	328.875	302.254	181.487	767.853	1324.461	
4	328.352	300.954	183.929	888.448	1343.424	

  

	WeakWashF	SteamHeatF-3	T-Top-Chips-4	SulphidityL-4
0	257.325	54.612	252.077	NaN
1	241.182	46.603	251.406	29.11
2	237.272	51.795	251.335	NaN
3	239.478	54.846	250.312	29.02
4	215.372	54.186	249.916	29.01

[5 rows x 23 columns]

```
In [21]: ► df_filled = df.fillna(3)
```

```
In [7]: print(df_filled)
```

	Observation	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	\
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	
..	...	...	...	...	...	...	
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	

  

	T-upperExt-2	T-lowerExt-2	UCZAA	WhiteFlow-4	...	SteamFlow-4	\
0	358.282	329.545	1.443	599.253	...	67.122	
1	351.050	329.067	1.549	537.201	...	60.012	
2	350.022	329.260	1.600	549.611	...	61.304	
3	350.938	331.142	1.604	623.362	...	68.496	
4	351.640	332.709	0.000	638.672	...	70.022	
..	...	...	...	...	...	...	
319	347.286	310.970	1.523	513.956	...	61.141	
320	399.135	319.576	1.451	570.058	...	67.667	
321	373.633	314.591	1.457	549.306	...	66.446	
322	364.081	308.559	1.523	504.852	...	61.054	
323	356.289	310.482	1.474	497.375	...	58.247	

  

	Lower-HeatT-3	Upper-HeatT-3	ChipMass-4	WeakLiquorF	BlackFlow-2	\
0	329.432	303.099	175.964	1127.197	1319.039	
1	330.823	304.879	163.202	665.975	1297.317	
2	329.140	303.383	164.013	677.534	1327.072	
3	328.875	302.254	181.487	767.853	1324.461	
4	328.352	300.954	183.929	888.448	1343.424	
..	...	...	...	...	...	
319	330.117	304.006	148.174	1027.201	1357.271	
320	330.848	304.616	165.178	906.962	1311.177	
321	330.226	304.686	160.841	887.125	1319.226	
322	327.346	304.363	147.589	804.423	1320.225	
323	328.092	304.093	144.218	828.328	1320.848	

  

	WeakWashF	SteamHeatF-3	T-Top-Chips-4	SulphidityL-4
0	257.325	54.612	252.077	0.00
1	241.182	46.603	251.406	29.11
2	237.272	51.795	251.335	0.00
3	239.478	54.846	250.312	29.02
4	215.372	54.186	249.916	29.01
..	...	...	...	...
319	381.643	45.264	252.947	30.86
320	25.494	50.528	252.092	30.70
321	0.638	45.549	252.438	0.00
322	0.000	43.725	253.176	31.13
323	1.276	43.840	253.216	0.00

[324 rows x 23 columns]

```
In [8]: df_dropped = df.dropna()
print(df_dropped)
```

	Observation	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	\
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	
5	1-08:00	14.23	15.350	85.518	1171.604	198.538	
7	31-06:00	22.65	14.100	91.887	1307.852	288.989	
9	31-08:00	24.70	13.850	96.208	1334.892	362.511	
..	...	...	...	...	...	...	
312	31-10:00	24.40	14.117	85.998	1330.104	394.234	
317	4-16:00	17.80	16.625	78.367	1276.082	202.744	
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	

  

	T-upperExt-2	T-lowerExt-2	UCZAA	WhiteFlow-4	...	SteamFlow-4	\
1	351.050	329.067	1.549	537.201	...	60.012	
3	350.938	331.142	1.604	623.362	...	68.496	
5	344.014	325.195	1.436	628.245	...	65.225	
7	352.321	331.162	1.468	625.549	...	71.298	
9	352.372	327.358	1.515	553.172	...	64.249	
..	...	...	...	...	...	...	
312	348.089	319.027	1.429	540.558	...	62.179	
317	360.127	329.266	1.488	698.486	...	75.296	
319	347.286	310.970	1.523	513.956	...	61.141	
320	399.135	319.576	1.451	570.058	...	67.667	
322	364.081	308.559	1.523	504.852	...	61.054	

  

	Lower-HeatT-3	Upper-HeatT-3	ChipMass-4	WeakLiquorF	BlackFlow-2	\
1	330.823	304.879	163.202	665.975	1297.317	
3	328.875	302.254	181.487	767.853	1324.461	
5	322.103	298.517	165.814	826.243	907.641	
7	329.662	301.539	179.886	837.178	1315.111	
9	332.264	305.419	166.120	909.810	1318.725	
..	...	...	...	...	...	
312	329.831	302.652	163.258	827.107	1312.372	
317	321.658	297.088	180.438	1017.333	1052.785	
319	330.117	304.006	148.174	1027.201	1357.271	
320	330.848	304.616	165.178	906.962	1311.177	
322	327.346	304.363	147.589	804.423	1320.225	

  

	WeakWashF	SteamHeatF-3	T-Top-Chips-4	SulphidityL-4
1	241.182	46.603	251.406	29.110
3	239.478	54.846	250.312	29.020
5	595.875	52.807	249.580	30.340
7	234.047	53.805	249.971	29.220
9	180.375	48.842	251.121	29.210
..	...	...	...	...
312	132.163	49.446	251.406	29.292
317	433.089	54.675	251.747	30.320
319	381.643	45.264	252.947	30.860
320	25.494	50.528	252.092	30.700
322	0.000	43.725	253.176	31.130

[141 rows x 23 columns]

```
In [9]: summary_stats = df.describe()
print("\nSummary statistics:")
print(summary_stats)
```

Summary statistics:

	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	\
count	324.000000	319.000000	307.000000	308.000000	323.000000	
mean	20.635370	14.347937	87.464456	1237.837614	258.164483	
std	3.070036	1.499095	7.995012	100.593735	87.987452	
min	12.170000	9.983000	68.645000	0.000000	0.000000	
25%	18.382500	13.358000	81.823000	1193.215250	213.527000	
50%	20.845000	14.308000	86.739000	1273.138500	271.792000	
75%	23.032500	15.517000	92.372000	1289.196000	321.680000	
max	27.600000	16.958000	121.717000	1351.240000	419.014000	

  

	T-upperExt-2	T-lowerExt-2	UCZAA	WhiteFlow-4	AAWhiteSt-4	\
count	322.000000	322.000000	299.000000	323.000000	173.000000	
mean	356.904295	324.020180	1.492010	591.732260	6.140410	
std	9.209290	7.621402	0.105923	67.016351	0.081609	
min	339.168000	284.633000	1.182000	405.111000	5.890000	
25%	350.241250	321.420000	1.431500	540.989500	6.089000	
50%	356.843000	325.669000	1.498000	592.895000	6.135000	
75%	362.242250	329.175000	1.560500	639.480500	6.199000	
max	399.135000	337.012000	1.747000	731.394000	6.340000	

  

	...	SteamFlow-4	Lower-HeatT-3	Upper-HeatT-3	ChipMass-4	\
count	...	323.000000	322.000000	322.000000	323.000000	
mean	...	66.668285	325.567820	300.525699	162.222322	
std	...	5.708587	4.609862	4.568484	14.160688	
min	...	48.568000	318.051000	293.312000	113.922000	
25%	...	62.518000	321.385500	296.513250	153.032500	
50%	...	67.429000	324.741000	299.126000	163.690000	
75%	...	71.522000	329.845250	304.244750	172.555000	
max	...	76.147000	333.854000	311.146000	189.268000	

  

	WeakLiquorF	BlackFlow-2	WeakWashF	SteamHeatF-3	T-Top-Chips-4	\
count	323.000000	322.000000	323.000000	322.000000	323.000000	
mean	873.828941	1175.917016	263.543068	49.696907	251.240087	
std	122.073521	149.334010	163.666942	4.551909	1.283432	
min	486.938000	838.948000	0.000000	35.510000	248.359000	
25%	792.019500	1044.817500	134.649000	46.389750	250.312000	
50%	865.254000	1150.221500	269.193000	50.277000	251.380000	
75%	965.286500	1319.021250	405.563000	53.294250	252.323500	
max	1226.277000	1395.767000	715.715000	63.332000	254.122000	

  

	SulphidityL-4
count	173.000000
mean	30.411671
std	0.701317
min	29.010000
25%	29.970000
50%	30.370000
75%	30.820000
max	32.840000

[8 rows x 22 columns]

```
In [25]: ► column_stats = df['BlowFlow'].describe()
print("\nSummary statistics for a specific column:")
print(column_stats)
```

Summary statistics for a specific column:

count 308.000000

mean 1237.837614

std 100.593735

min 0.000000

25% 1193.215250

50% 1273.138500

75% 1289.196000

max 1351.240000

Name: BlowFlow, dtype: float64

```
In [ ]: ►
```